

# Resource Management in the Tessellation Manycore OS \*

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## Abstract

Tessellation is a manycore OS targeted at the resource management challenges of emerging client devices, including the need for real-time and QoS guarantees. It is predicated on two central ideas: *Space-Time Partitioning* (STP) and *Two-Level Scheduling*. STP provides performance isolation and strong partitioning of resources among interacting software components, called *Cells*. Two-Level Scheduling separates *global* decisions about the allocation of resources *to* Cells from *application-specific* scheduling of resources *within* Cells. We describe Tessellation’s Cell model and its resource allocation architecture. We present results from an early prototype running on two different platforms including one with memory-bandwidth partitioning hardware.

## 1 Introduction

The trend toward *manycore* systems (with 64 or more cores) presents serious challenges for client devices. Users will expect better performance from applications as the number of cores increase; this expectation will be challenging to meet since it requires parallelizing client applications (which are often not very scalable) and exploiting parallelism that is likely to be fragile and easily disturbed by interference. Further, tomorrow’s applications will consist of variety of components – each of which presents complex and differing resource requirements. In addition to best-effort computation, users have an increasing appetite for responsive user interfaces and high-quality multimedia (e.g., multi-party videoconferencing, multi-player gaming, and music composition) with stringent real-time requirements; such needs are not well supported by today’s commodity operating systems.

We believe that the advent of manycore is an opportunity to fundamentally restructure operating systems to support a simultaneous mix of interactive, real-time, and high-throughput parallel applications. Our hypothesis is that a much wider variety of performance goals can be met by structuring the operating system around resource distribution, performance isolation, and QoS guarantees; such structuring is natural in a manycore environment.

This paper investigates the combination of two complementary ideas embodied in our new OS, called *Tessellation*: *Space-Time Partitioning* and *Two-Level Scheduling*. *Space-Time Partitioning* (STP) [21], exploits novel

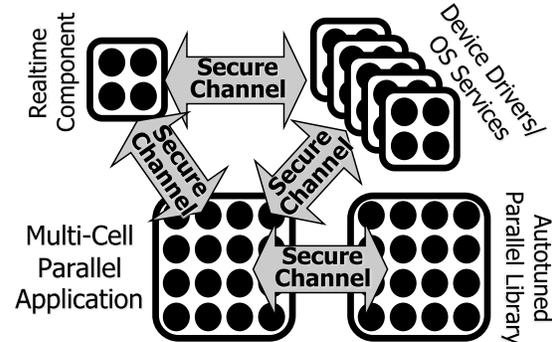


Figure 1: Decomposing an application into a set of communicating components and services running with QoS guarantees within Cells. Tessellation provides Cells that host device drivers and OS services.

software layering and hardware mechanisms (when available) to support a model of computation in which applications are divided into performance-isolated, gang-scheduled *Cells* communicating through secure channels; see Figure 1.

Two-Level Scheduling separates global decisions about the allocation of resources *to* Cells from application-specific scheduling of resources *within* Cells. The resource distribution process (first level) is one of the novel elements of our approach and is discussed in detail in Section 3. Once resources have been assigned to Cells, STP guarantees that user-level schedulers within Cells (second level) may utilize resources as they wish – without interference from other Cells or from the OS. It is the separation of resource distribution from usage that we believe makes Two-Level Scheduling more scalable than other approaches and better able to meet the demands of parallel client applications.

## 2 Overview of Tessellation

Tessellation is a manycore OS focused on resource guarantees. Here we summarize key aspects of Tessellation.

### 2.1 Space-Time Partitioning

A *spatial partition* (or *partition*) is a performance-isolated unit of resources maintained through a combination of software and hardware mechanisms. Managed resources include gang-scheduled hardware thread contexts, guaranteed fractions of shared resources (e.g., cache or memory bandwidth), access to OS services, and fractions of the energy budget. Tessellation divides the hardware into a set of simultaneously-resident partitions

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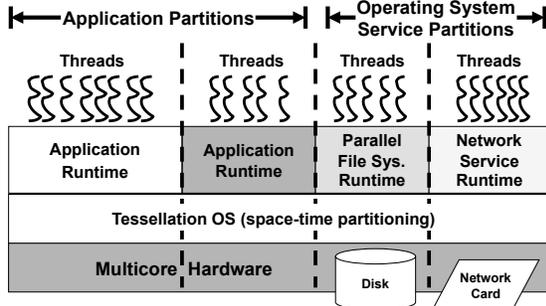


Figure 2: *Space-Time Partitioning* in Tessellation: a snapshot in time with four spatial partitions.

as shown in Figure 2. Partitioning varies with the needs of the OS and applications – hence the “time” component of the term Space-Time Partitioning (STP).

Support for STP consists of a combination of hardware and software mechanisms. The “Partition Mechanism Layer” of Tessellation, which enforces partition boundaries, has some similarities to a hypervisor [1, 4, 16] but with a crucial difference: its sole task is to provide performance-isolated, QoS-guaranteed containers for applications and OS services. Although Tessellation runs on existing multicore systems, it can also exploit hardware enhancements when available.

## 2.2 The Cell Model

Tessellation exports STP to applications and OS services through an abstraction called a *Cell*. A Cell is a container for parallel software components providing guaranteed access to resources, *i.e.*, the performance and behavior of an isolated machine. Resources are guaranteed as space-time quantities, such as “4 processors for 10% of the time” or “2 GB/sec of bandwidth”. Although Cells may be time-multiplexed, hardware thread contexts and resources are gang-scheduled such that Cells are unaware of this multiplexing. In other words, unexpected virtualization of physical resources does not occur.

Resources allocated to a Cell are owned by that Cell until explicitly revoked<sup>1</sup>. Once the Cell is mapped, a user-level scheduler is responsible for scheduling hardware contexts and other resources. There is no paging of physical memory unless a paging library is linked into the user-level runtime. Further, each Cell’s runtime has control over the delivery of events such as inter-cell messages, timer interrupts, exceptions, and faults.

**Inter-Cell Communication:** Inter-cell communication occurs through *channels*. A channel provides performance and security isolation between Cells. The setup and tear-down of a channel is privileged and strictly controlled by the OS. Once constructed, a channel provides fast asynchronous communication at user-level.

<sup>1</sup>Tessellation notifies the user-level scheduler of revocations, giving it a chance to adjust accordingly.

**Utilizing Cells for OS Services:** Cells provide a convenient abstraction for building OS services such as device drivers, network interfaces, and file systems. Tessellation adopts a philosophy similar to that of microkernels [8]. Unlike traditional microkernels, however, multiple components can be mapped to the hardware simultaneously – allowing rapid inter-domain communication. Further, each interacting component is explicitly parallel and performance-isolated from other components.

Partitioning OS functionality into a set of interacting Cells provides predictable and reliable behavior due to limited interaction with the rest of the system. QoS guarantees on shared services can be enforced by restricting channel communication. Alternatively, the capacity of overloaded services can be increased by resizing Cells.

## 2.3 Two-level Scheduling in Tessellation

Tessellation separates global decisions about resource allocation from local decisions about resource usage. The result is a good match to the abundant resources present in a manycore system: The resource allocation process can focus on the impact of resource *quantities* on Cell execution – leaving the fine details about how to utilize these resources to application-specific schedulers.

**Distributing Resources to Cells:** The resource allocator, described in Section 3, distributes partitionable resources among Cells and exercises the option to reserve or deactivate resources to guarantee future responsiveness or to optimize energy consumption. Tessellation changes resource distribution infrequently to amortize the cost of the decision-making process and to minimize interference with application-level scheduling.

**Scheduling Within a Cell:** The Cell-level scheduler runs at user-level and manages all resources within the Cell. Performance isolation between Cells guarantees that applications can get predictable and repeatable behavior, simplifying performance optimization and real-time scheduling. Central to Tessellation’s approach are runtime frameworks, such as Lithe [23], that produce composable, application-specific schedulers. Via Lithe, Tessellation supports a variety of parallel programming models in a uniform and composable way.

## 3 Resource-Allocation Architecture

Tessellation strikes a balance between maximizing resource utilization to achieve performance and selectively idling resources to provide QoS guarantees. Decision-making logic is packaged into a *Policy Service* that distributes resources to Cells by combining system-wide goals, resource constraints, and performance targets with current performance measurements (see Figure 3). The results are passed as a *Space-Time Resource Graph* (STRG) to the Tessellation kernel for QoS enforcement.



figurations of Cells nor is it capable of incorporating subtleties in the level of importance of some Cells over others. Instead, we believe that two components are necessary: first, accurate models of the performance of Cells (or the change in performance) as a function of resources and second, a framework in which to drive the juggling of resources among Cells. One framework that we are investigating is a form of convex optimization over resources that attempts to minimize an “urgency” derived from the degree to which Cells miss their deadlines [26].

**Modeling Application Behavior:** The process of determining how client application performance varies with low-level resources is often labor-intensive and error-prone. Except in very special circumstances, programmers are unlikely to know “exactly” how low-level resources affect their performance goals. For example, the programmer may want a given *frame rate* but have no idea how much memory bandwidth is required to meet that rate. This conundrum is a form of “impedance mismatch” between the units of hardware resources and the programmer-specified QoS requirements [17].

It is therefore desirable to figure out how to *automatically* construct predictive models of application performance. Tessellation’s performance isolation should permit us to build highly accurate models. However, we must decide when and how these models will be built.

One solution is to profile the applications in advance. This solution could be viable for application distribution models like the iTunes Application Store or the Android Market. The application distributor can profile the applications for the limited platforms it supports and provide predictive models when the application is downloaded.

Another option leverages the Cloud by requesting every user to record performance, resource, and platform statistics. This approach is currently being used by Microsoft Research to generate performance models of Microsoft applications for developers [13].

A more general solution is to have the operating system profile the applications online and use the locally collected information to make resource decisions. While the operating system could potentially run the applications over a set of configurations when the application is first installed to build the models, it may be more practical to simply start with a generic model which is refined over time with on-line training methods.

The source of the data used to create the models is somewhat orthogonal to the type of models used. Adopting a particular model type in turn influences the specific method used for searching over all models to find the best resource distribution across all Cells. We are currently interested in two types of models. Models of the first type are based on a discrete, sparse set of predetermined operational points (*e.g.*, [12]). Models of the

second type use continuous functions to capture performance trade-offs across a large number of allocations (*e.g.*, [5, 9]). The effectiveness of linear and quadratic models has been shown in [7].

### 3.3 Partition Mapping and Multiplexing

The *Partition Mapping and Multiplexing Layer* (or *Mapping Layer*), translates the resource specifications of the Policy Service (expressed in the STRG) into an ordered sequence of spatial partitions for the underlying *Partition Mechanism Layer* (mentioned in Section 2). The Mapping Layer makes no policy decisions, but rather implements the policy decisions given by the Policy Service.

The Mapping Layer comprises two main components: the *Planner* and the *Plan Executor*. When the Planner receives a new STRG from the Policy Service, it first validates that this STRG does not violate basic security or QoS requirements<sup>4</sup>, then generates a future plan for distributing resources to Cells. The Planner invokes an operation similar to bin-packing to assign Cells and resources to future partition time-slices.

The Plan Executor implements the resulting resource plan. It can modify the plan being executed in predefined ways to accommodate more dynamic resource-allocation and time-multiplexing actions (*e.g.*, activation of a Cell upon the arrival of an event or redistribution of excess resources among Cells).

In implementing the STRG, the Mapping Layer implements a variety of *Cell activation policies*. Examples include the *Pinned Policy* (Cell given dedicated access to cores), the *Time-Triggered Policy* (Cell active during predetermined time-windows for real-time predictability), and the *Time-fraction Policy* (Cell active for a specified fraction of the time). Most Cell activation policies are *non-preemptive*: once a Cell is activated it is not suspended until its time-slice expires. The one exception is that Cells can be given best-effort resources that may be preempted by Cells with higher priority.

## 4 Experimental Evaluation

In this section, we examine the potential for performance isolation in the Tessellation prototype. The prototype was derived from an early version of the ROS kernel [19], supplemented with support for cell time-multiplexing and second-level preemptive scheduling. This prototype contains 22,000+ lines of code and runs on both Intel x86 platforms and RAMP Gold [27, 28]. RAMP Gold is an FPGA-based simulator that models up to 64 in-order 1-GHz SPARC V8 cores, a shared memory hierarchy, and hardware partitioning mechanisms. The Intel system used in our experiments is equipped with dual 2.67-GHz Xeon X5550 quad-core processors.

<sup>4</sup>We are exploring how to remove as much of the Policy Service from the trusted computing base as possible.

	2 Cores	15 Cores	63 Cores
Intel activate	1.57 $\mu$ s	8.26 $\mu$ s	N/A
RAMP activate	0.69 $\mu$ s	1.88 $\mu$ s	5.37 $\mu$ s
Intel suspend	1.58 $\mu$ s	17.59 $\mu$ s	N/A
RAMP suspend	1.19 $\mu$ s	5.91 $\mu$ s	34.10 $\mu$ s

Table 1: Mean activation and suspension latencies for cells of varying size. Here, core 0 was dedicated to Cell management.

**Cell Activation and Suspension:** Table 1 summarizes the overhead of activating and suspending a Cell with varying core counts on both RAMP Gold and our Intel system. These numbers are preliminary. The overhead is small relative to the time scale of Cell time-multiplexing (e.g., 100 ms), but is still larger than we would like.

**Performance Isolation:** Our 64-core RAMP Gold platform simulates a mechanism that can dedicate fractions of off-chip memory bandwidth to Cells [20]. We illustrate Tessellation’s use of this mechanism by creating three Cells as follows:  $Cell_1$  is given 32 cores and 50% of memory bandwidth (i.e., 6.4GB/s);  $Cell_2$  16 and 25%; and  $Cell_3$  15 and 25%.  $Cell_1$  contains the PARSEC *streamcluster* [6], selected for its significant memory capacity and bandwidth requirements. Other PARSEC benchmarks run in the remaining Cells.

We first activate  $Cell_1$  by itself. Next, we activate all Cells and run the benchmarks concurrently both with and without memory bandwidth partitioning.  $Cell_1$  takes 5.70M, 6.12M and 11.59M core-cycles on average to complete, for the three experiments. The respective standard deviations are 0.30M, 0.95M and 1.17M. These results show that Tessellation can provide significant performance isolation.

**Spatial Partitioning:** Here we evaluate the potential of spatial partitioning using Tessellation on the RAMP Gold 64-core machine. We take pairs of PARSEC applications, placing each application in a Cell. Cores are assigned in groups of 8, page colors in sets of 16, and memory bandwidth in units of 3.4GB/s (a combinatorial total of 54 valid allocations). We then evaluate *all* possible spatial allocations for the two Cells. We also evaluate the case in which one Cell is assigned the entire machine and run to completion followed by the other Cell; this is a favorable time-multiplexing scenario as there is no overhead from repeated context switches.

For some pairs, time-multiplexing is better than any possible spatial-partitioning. However, for many pairs, the optimal spatial partition is substantially better thanks to disjoint resource requirements, reduced interference between pairs, or imperfect application scaling. Figure 4 illustrates the performance of several pairs. It shows that spatial partitioning can provide significant performance benefits. However, naïve spatial divisions are likely to be detrimental to performance, meaning that the Policy Service must be judicious in assigning resources to Cells.

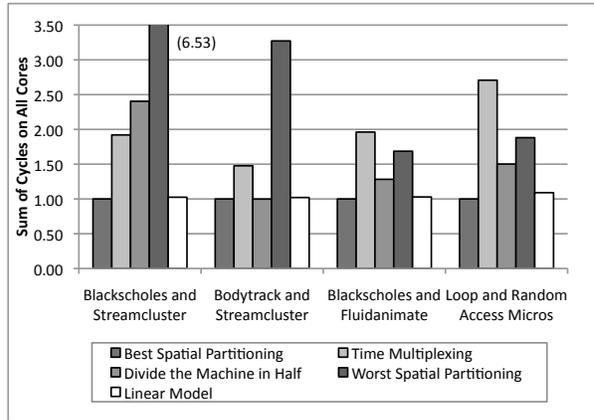


Figure 4: Performance of spatial partitioning compared with a favorable time-multiplexing. Performance is in cycles (lower is better) and results are normalized to the best spatial partition.

We also found that a *simple* linear approximation model, derived from only 10 sample points, allowed us to make resource-allocation decisions within 10% of optimal every time. These results encourage further research on approaches to resource allocation for Cells and tradeoffs between complexity, overhead and performance.

## 5 Related Work

Tessellation is influenced by virtual machines, exokernels, and multiprocessor runtime systems [1, 2, 4, 10, 15, 16, 18, 24]. Other recent manycore operating systems projects, such as Corey [11], Barrelfish [25], and fos [29], share some structural aspects such as distributed OS services. This body of work mainly focus on improving OS scalability and, contrary to Tessellation, does not attempt to also provide QoS guarantees.

Nesbit et al. [22] introduce Virtual Private Machines (VPM), another framework for resource allocation and management in multicore systems. The concepts of VPM and Cell are similar, but the VPM framework does not include an equivalent communication mechanism to our inter-cell channel.

## 6 Conclusion

We presented Tessellation, a new manycore OS for client devices that provides real-time and QoS guarantees. Tessellation is predicated on two central ideas, namely *Space-Time Partitioning* (STP) and *Two-Level Scheduling*. In this paper, we discussed Tessellation’s Cell model, explored its resource-allocation architecture, and examined results from an early prototype.

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